

## IV. Capture-Recapture Models with Individual Covariates



# INDIVIDUAL COVARIATE MODELS

- $p$  depends on an explicit covariate which you only observe the values of for individuals that are encountered
- $\text{logit}(p[i]) = \alpha_0 + \alpha_1 x[i]$
- For this reason, called “**Model M<sub>x</sub>**” (Kery and Schaub BPA)
- Two diametrically opposite approaches to analysis
  - (1) Horvitz-Thompson estimation (Huggins and Alho used this idea based on conditional likelihood)
  - (2) Model-based “full likelihood”: Put a distribution on  $x[i]$  (Borchers et al. 1998; Royle 2009, Biometrics)

# Model M<sub>x</sub>

We still have a binomial encounter model:

$$y[i] \sim \text{Binomial}(p[i], K)$$

$$\text{logit}(p[i]) = a + b \cdot x[i]$$

- $x[i]$  is NOT OBSERVED for uncaptured individuals

Model for the covariate:

$$x[i] \sim \text{normal}(\mu_x, \sigma_x) \text{ (or similar)}$$

# INDIVIDUAL COVARIATES

- Huggins, R. M. (1989). On the statistical analysis of capture experiments. *Biometrika*, 76(1), 133-140.
- Alho, J. M. (1990). Logistic regression in capture-recapture models. *Biometrics*, 46(3), 623.
- Borchers, D. L., Zucchini, W., & Fewster, R. M. (1998). Mark-recapture models for line transect surveys. *Biometrics*, 1207-1220.
- Pollock, K. H. (2002). The use of auxiliary variables in capture-recapture modelling: an overview. *Journal of Applied Statistics*, 29(1-4), 85-102.
- Royle, J. A. (2009). Analysis of capture-recapture models with individual covariates using data augmentation. *Biometrics*, 65(1), 267-274.
- Borchers, D.L., S.T. Buckland and W. Zucchini. 2002. Estimating Animal Abundance: Closed Populations (Springer Verlag) (Ch. 11).

# Model Mx and multi-session models

- Conceptually and technically Model Mx is exactly like “class-structured” models considered previously except Model Mx usually used in context of a continuous covariate.

- Consider having sex-specificity of model parameters

$$\text{logit}(p[i]) = \alpha_0 + \alpha_1 * X_{\text{sex}}[i]$$

$$X_{\text{sex}}[i] = \begin{array}{l} 0 \text{ if female} \\ 1 \text{ if male} \end{array}$$

- $X_{\text{sex}}$  is missing for M-n individuals in our augmented data set.
- Put a prior distribution on it....With a discrete covariate, the prior is “class membership”

# Example of model Mh

- Microtus data from Williams et al. (2002)

```
source("microtus.data.R")  
head(microtus.data)
```

	[, 1]	[, 2]	[, 3]	[, 4]	[, 5]	[, 6]
[1, ]	0	1	1	0	0	37
[2, ]	1	0	1	1	0	46
[3, ]	1	1	1	0	1	60
[4, ]	1	1	1	1	1	49
[5, ]	0	0	0	0	1	38
[6, ]	1	0	1	0	1	40

$n = 56$        $K = 5$  sample occasions ,  $x[i]$  = "body mass" is stored in column 6 of the matrix

Model for  $x[i]$ :  $x[i] \sim \text{Normal}(\mu, \sigma)$

# Analysis of the Microtus data

- R work session

R script: `closed_models_part4.R`

# Model Mx: toward SCR

Model Mx has been widely adopted for estimation of N in capture-recapture studies to account for spatial heterogeneity in encounter probability, by defining:

$x$  = “distance to edge” (DTE)

This is estimated for each captured individual and treated as fixed and known.

Boulanger and McLellen (2001)

Ivan and White (2013)

**Problems:** variable precision. Biased near edge! Also doesn't account for trap-level information.

# Applying model M<sub>x</sub>

- We'll use  $x$  = distance to centroid (DTC) of trap array, call this "**xdist**":

$$\text{xdist}[i] = \text{dist}(\text{sbar}[i], x_0)$$

$x_0$  = mean trap location (centroid of trap array)

$\text{sbar}[i]$  = average location of individual  $i$ .

To do the Bayesian analysis by DA we need a prior for  $\text{xdist}[i]$  to account for uncaptured individuals. **Could as well just put the prior on  $\text{sbar}[i]$  since  $x_0$  is known.**

# Prior for $d[i]$ or prior for $s[i]$ ?

- **Prior for  $x_{\text{dist}}[i]$ :** What are the possible values for where captureable individuals come from?
- $x_{\text{dist}}[i] \sim \text{dunif}(0, D_{\text{max}})$

$D_{\text{max}}$  = furthest possible capture?

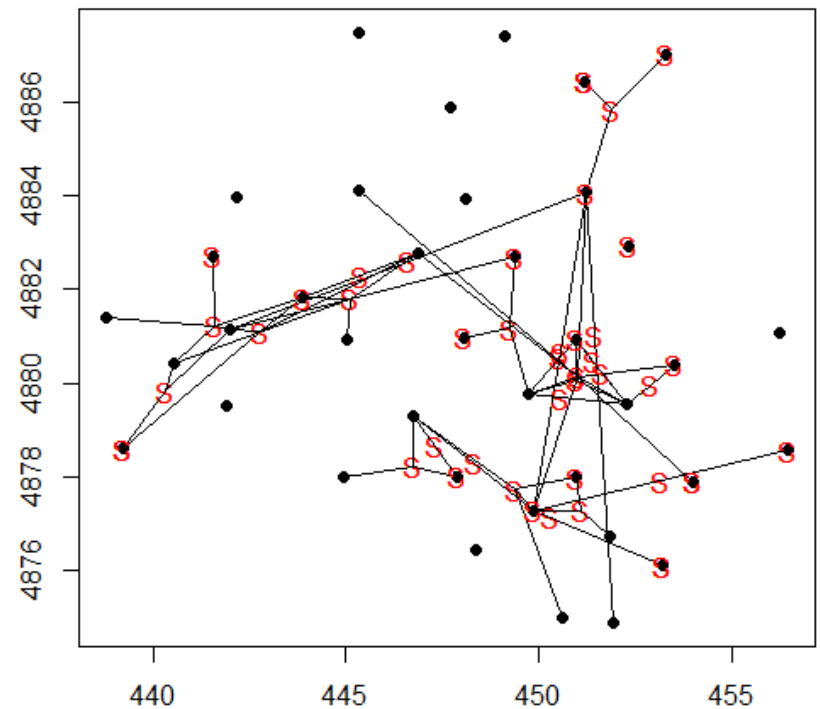
# Fort Drum black bear data

```
library("scrbook")
data(beardata)
nind<-dim(beardata$bearArray)[1]
K<-dim(beardata$bearArray)[3]
ntraps<-dim(beardata$bearArray)[2]
toad<- spiderplot(beardata$bearAr

## Distance to centroid of traps
xdist<-toad$xcen

## average location of capture
sbar<- toad$avg.s

## Centroid of trap array
x0<- toad$center
```



# ESTIMATING DENSITY

- By putting a prior distribution on  $x_{dist}[i]$  this explicitly defines an AREA within which the sampleable population lives. That is,  $N$  is all individuals within  $D_{max}$  of the centroid
- You will find that the estimated  $N$  changes as you change  $D_{max}$ .
- The DTC/DTE model, with a prior on the distance covariate, simultaneously estimates  $N$  and Density
- Provides resolution to the “unknown area” problem
- (was **not** noted by Boulanger and McLellan 2001)

# MODEL MX HAS PROBLEMS

(1) Subjective choice of  $D_{max}$  has a big effect – this model implies that density of individuals decreases as you move away from the centroid

Area of concentric rings INCREASES as you move away. So a constant frequency of individuals corresponds to lower density

(2) Use of estimated DTC (or DTE) is biased and estimated with variable precision. Model does not account for that.

# IMPROVING MODEL MX

- **Improvement 1:** Instead of messing with a prior on  $x_{\text{dist}}[i]$  why not just put the prior on  $s_{\text{bar}}[i]$ ? Exactly the same model, just a different prior (via a transformation).
- **Improvement 2:** Instead of distance to  $x_0$ , why not distance to each trap  $x[j]$ ?

$$x_{\text{dist}}[i,j] = \text{dist}(s_{\text{bar}}[i], x[j])$$

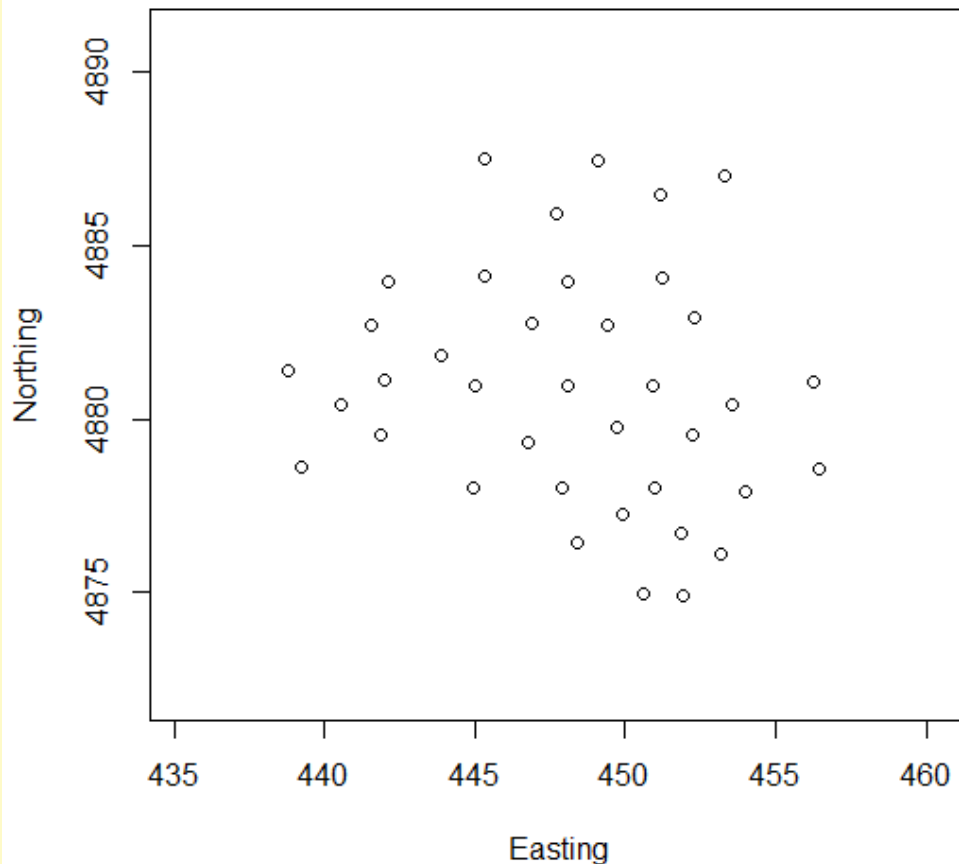
- **Improvement 3:** “ $s_{\text{bar}}$ ” is really a surrogate for “center of activity” – which is unobserved. Make it a latent variable (like Model Mh but with some indirect information)

# Improvement 1 of model Mx

Improvement 1: Instead of messing with a prior on  $x_{dist}[i]$  why not just put the prior on  $sbar[i]$ ? **Exactly the same model, just a different prior (via a reparameterization).**

- $sbar[i]$  is the average capture location. But it's really a surrogate for “where individual  $i$  lives” – home range center?
- What kind of prior makes sense for this? In the absence of information, how about  $sbar[i] \sim \text{Uniform}(\text{space around trap array})$  ???  
S = “space around the trap array”

# The prior distribution for “sbar” for the Fort Drum model



- sbar is Uniform(S)
- S defined by: 4 unit buffer around the minimum and maximum x- and y-coordinates
- Try different buffers and verify stabilizing Density =  $N/\text{area}$

# A WARNING ABOUT COORDINATE SCALING IN BUGS

- The model which regards sbar as a variable effectively predicts sbar for each **uncaptured individual**
- WinBUGS seems to only carry around 4 significant digits (or else R2WinBUGS rounds to 4 digits, we're not sure).
- The coordinates of sbar for Fort Drum is 4xx.x and 48x.x – the leading 4 and 48 are costing precision for estimating the coordinates of uncaptured bears due to this 4-digit truncation
- JAGS does not appear to suffer this problem
- If we use BUGS it is imperative that we scale/translate the coordinate system so that we're not carrying around unnecessary digits (or use JAGS)

# Improvement 2 of model Mx

**Improvement 2:** Instead of distance to centroid, why not distance to each trap  $x[j]$ ?

$$x_{\text{dist}}[i,j] = \text{dist}(\text{sbar}[i], x[j])$$

$$\text{logit}(p[i,j]) = \alpha_0 + \alpha_1 * x_{\text{dist}}[i,j]$$

Note:  $p$  now depends on  $i$  **and**  $j$

Traps are just replicate sample occasions, like distinct methods, or sample frames, or observers, but with trap-specific  $p$ .

# Improvement 3 of model $M_x$

- $\bar{s}$  is meant to be an estimate of something, say  $s$ , the centroid of activity – “activity center”, home range center, etc..
- $s$  is strictly unknown. Regard it as a latent variable.
- In BUGS: input initial values for it, remove from “data”
- No plug-in estimation bias and heterogeneous variance.

# SUMMARY OF CAPTURE-RECAPTURE PART 4

- Individual covariate models with distance-to-edge/distance-to-centroid resolve some technical problems with ordinary CR models
  - Heterogeneity in  $p$  related to variable exposure to trapping
  - Putting a distribution on the covariate resolve “unknown area”
- Useful as a starting point for developing fully spatial capture-recapture models
  - Model location instead of distance
  - Distance to each trap
  - Treat “ $s$ ” as a latent variable